

Application of DNB for air quality and fire monitoring



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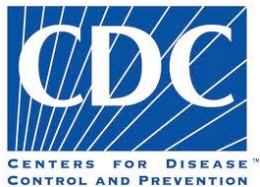


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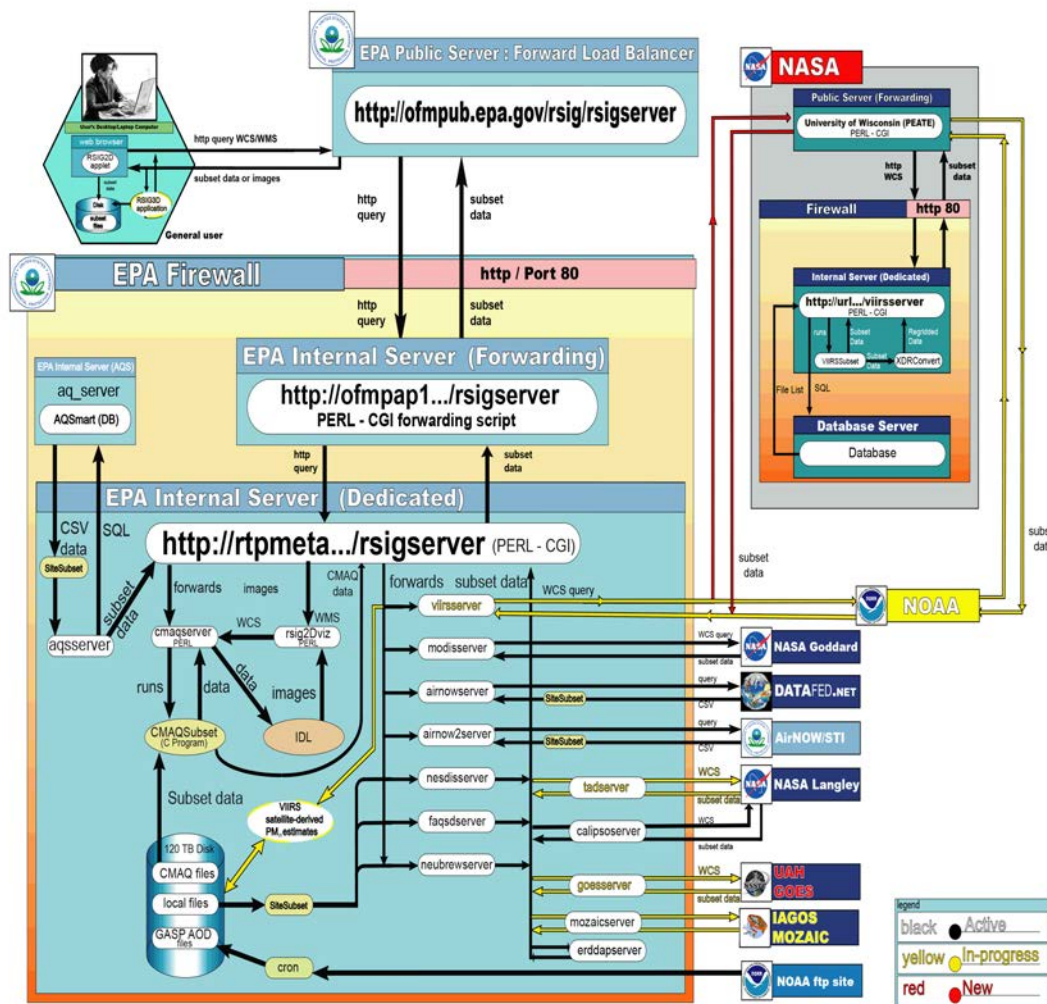


Robert Holz



Use of EPA Remote Sensing Information Gateway to deliver VIIRS AOD/PM_{2.5} data products

RSIG Data Flow Diagram



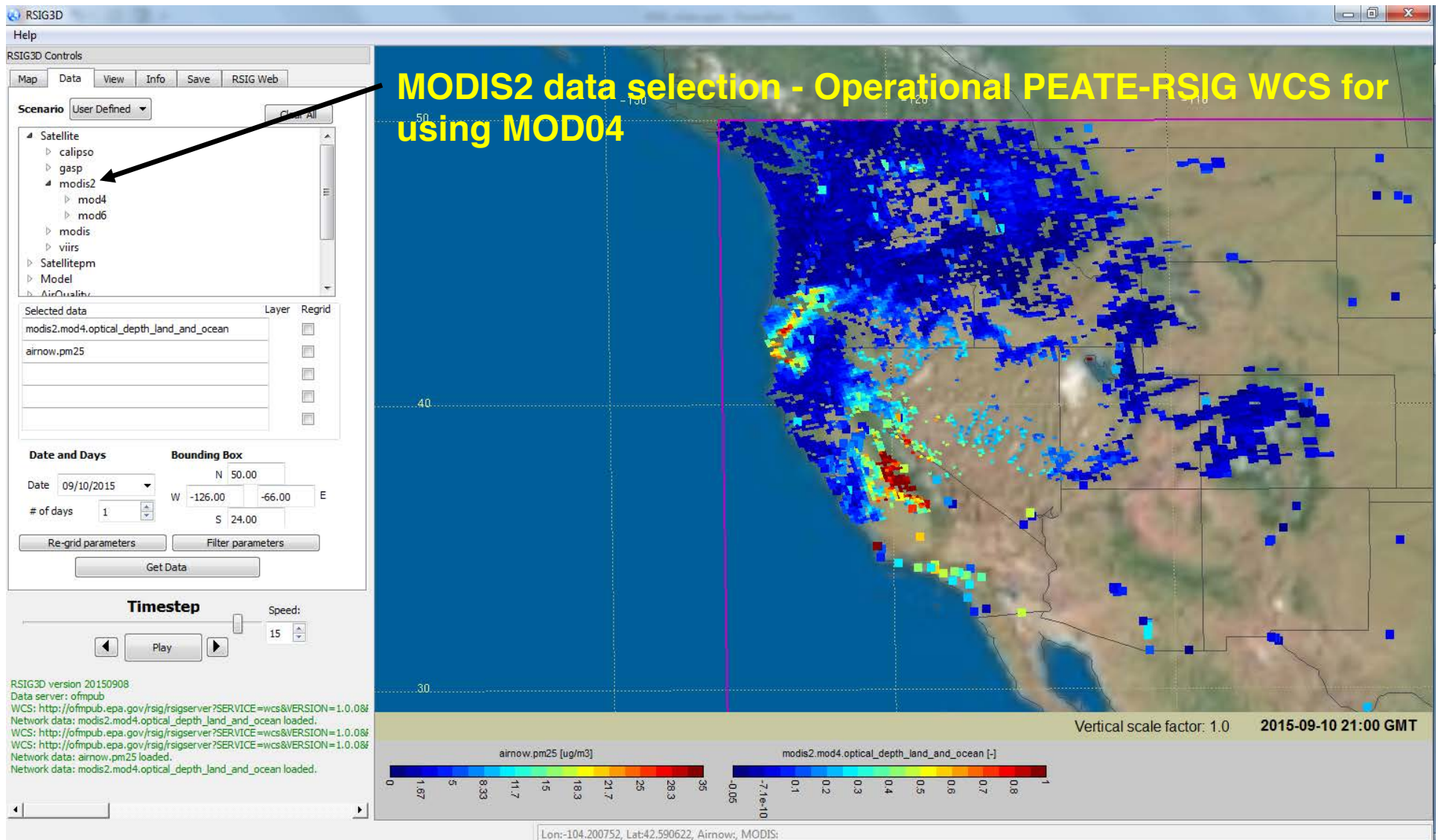
<http://ofmpub.epa.gov/rsig/rsigserver?index.html>

- Current satellite WCS:
 - MODIS C6 (10 km, 3 km, DB)
 - CALIOP, GASP (GOES AOD)
 - Prototype NOAA-VIIRS
- Establish OGC compliant Web Coverage Service (WCS) between PEATE and RSIG to add NASA- VIIRS data (This project). --- Done !!!
- GEOS-Chem scaling factors used to create a daily Look-Up-Table (LUT) of the spatial varying relation of AOD and PM_{2.5} (van Donkelaar et. al., 2012, ES&T) .
- Prototype use of AOD-to-PM2.5 scaling factors via regional models (WRF-CMAQ & WRF-CHEM) and explore ensemble type approach (This project).

Summary

- Data flow from UW-SIPS (Science Investigator Processing System) to EPA's RSIG is implemented, tested, and successful. [ARL4->ARL-7](#)
- Evaluation of ensemble approach for surface PM_{2.5} estimates from VIIRS and other satellite projects is conducted for June 2012. This would provide insight on the selection and improvement of operation approach for remote sensing of surface PM_{2.5}. [ARL2->ARL3](#)
- Refinement of Hierarchical Autoregressive Model has started in EPA site. [ARL5->ARL7](#)

Developed and implemented operational PEATE-RSIG WCS using MOD04; to be switched to VIIRS data products when final file format implemented on PEATE





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Potential application of VIIRS Day/Night Band for monitoring nighttime surface PM_{2.5} air quality from space

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H I G H L I G H T S

- VIIRS Day/Night Band (DNB) is much more sensitive to aerosols than to water vapor
- Modeling of outdoor light transfer in nighttime atmosphere for VIIRS DNB
- DNB potential for estimating surface PM_{2.5} is shown qualitatively and quantitatively
- PM_{2.5} at VIIRS night overpass time is much closer to daily-mean PM_{2.5} than at daytime
- Strategies for future DNB remote sensing of aerosols are elaborated

National Ambient Air Quality Standards

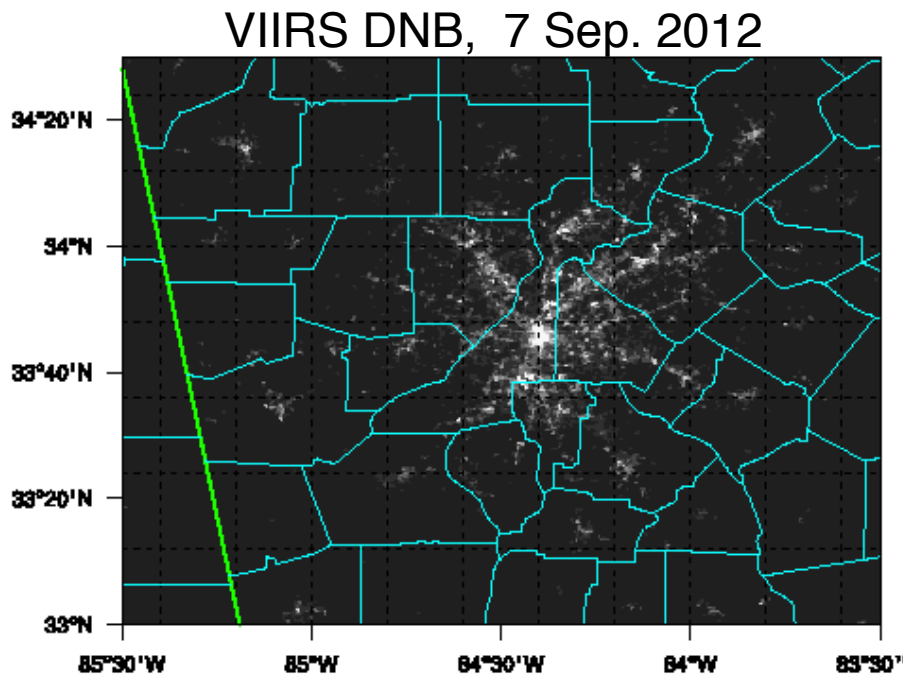
Pollutant [final rule cite]		Primary/ Secondary	Averaging Time	Level
<u>Carbon Monoxide</u> [76 FR 54294, Aug 31, 2011]		primary	8-hour	9 ppm
			1-hour	35 ppm
<u>Lead</u> [73 FR 66964, Nov 12, 2008]		primary and secondary	Rolling 3 month average	0.15 µg/m ³ ⁽¹⁾
<u>Nitrogen Dioxide</u> [75 FR 6474, Feb 9, 2010] [61 FR 52852, Oct 8, 1996]		primary	1-hour	100 ppb
		primary and secondary		12 µg/m ³ , FR, 15 Jan. 2013 ⁽²⁾
<u>Ozone</u> [73 FR 16436, Mar 27, 2008]		primary and secondary	8-hour	0.075 ppm ⁽³⁾
<u>Particle Pollution</u> [71 FR 61144, Oct 17, 2006]	PM _{2.5}	primary and secondary	Annual	15 µg/m ³
			24-hour	35 µg/m ³
	PM ₁₀	primary and secondary	24-hour	150 µg/m ³
<u>Sulfur Dioxide</u> [75 FR 35520, Jun 22, 2010] [38 FR 25678, Sept 14, 1973]		primary	1-hour	75 ppb ⁽⁴⁾
		secondary	3-hour	0.5 ppm

NAAQS uses **daily** and annual averages of $\text{PM}_{2.5}$

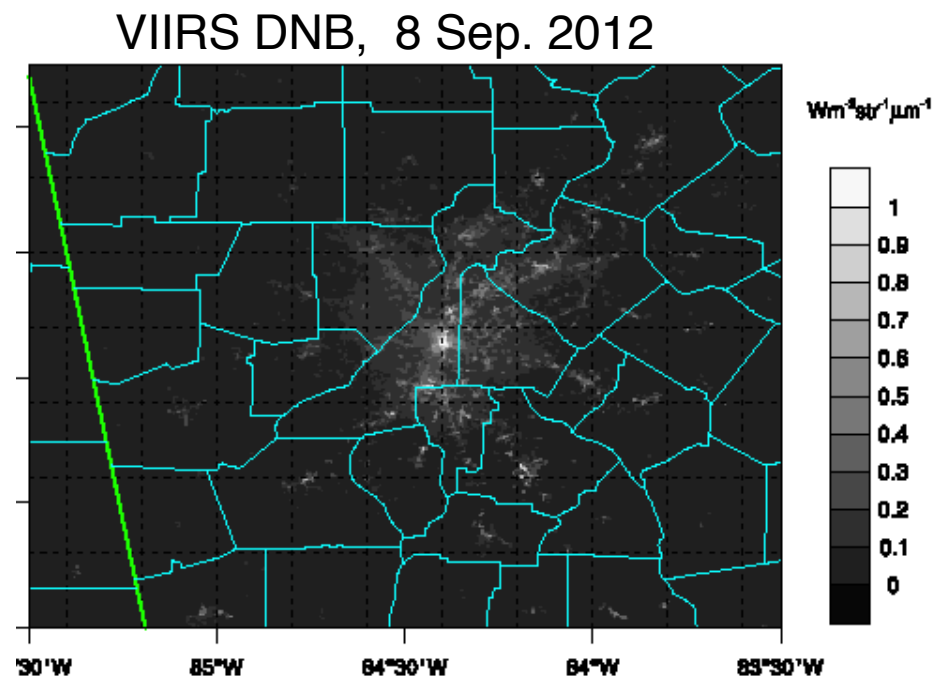
Can we use DNB to estimate surface $\text{PM}_{2.5}$ at night?

- At night, aerosols are often mixed in a shallow nocturnal boundary layer.
- Retrieval of AOD from DNB is still in its infancy; preliminary work include Zhang et al. (2008) and Johnson et al. (2013).
- We like to make a first attempt to apply DNB for night time $\text{PM}_{2.5}$ air quality.
- Aug – Oct 2012. Focus area: Atlanta

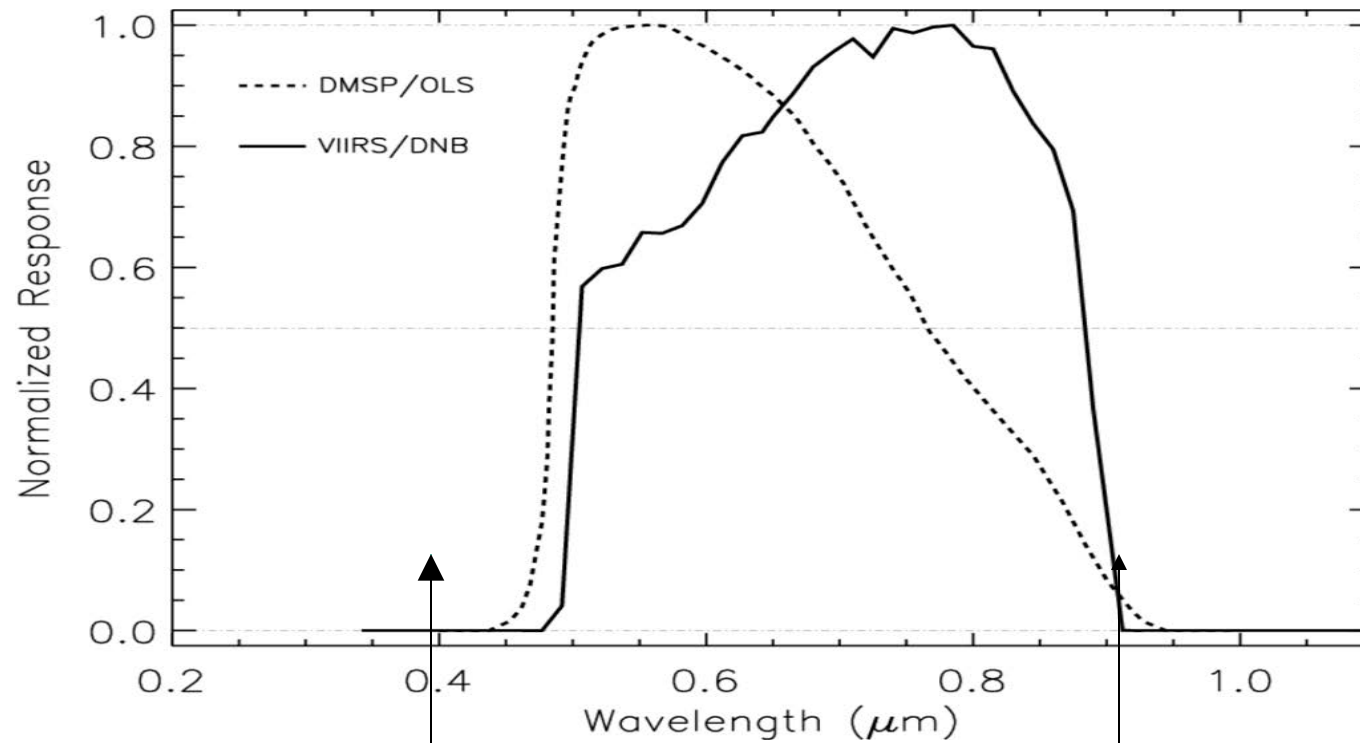
$\text{PM}_{2.5}$: 5 $\mu\text{g}/\text{m}^3$



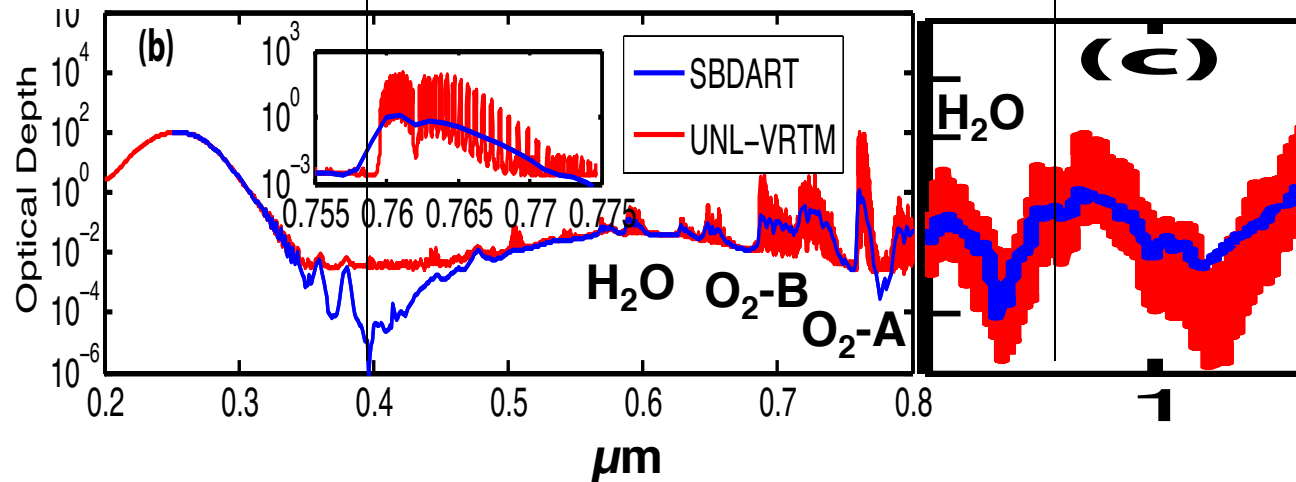
$\text{PM}_{2.5}$: 13 $\mu\text{g}/\text{m}^3$



Is DNB sensitive to aerosol, water vapor, & O₂ absorption ?

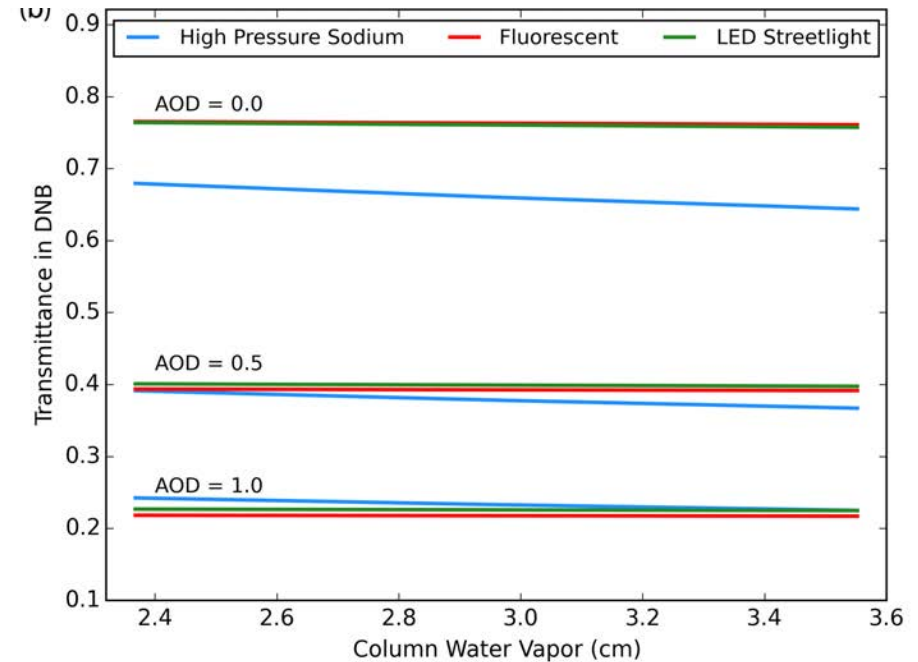
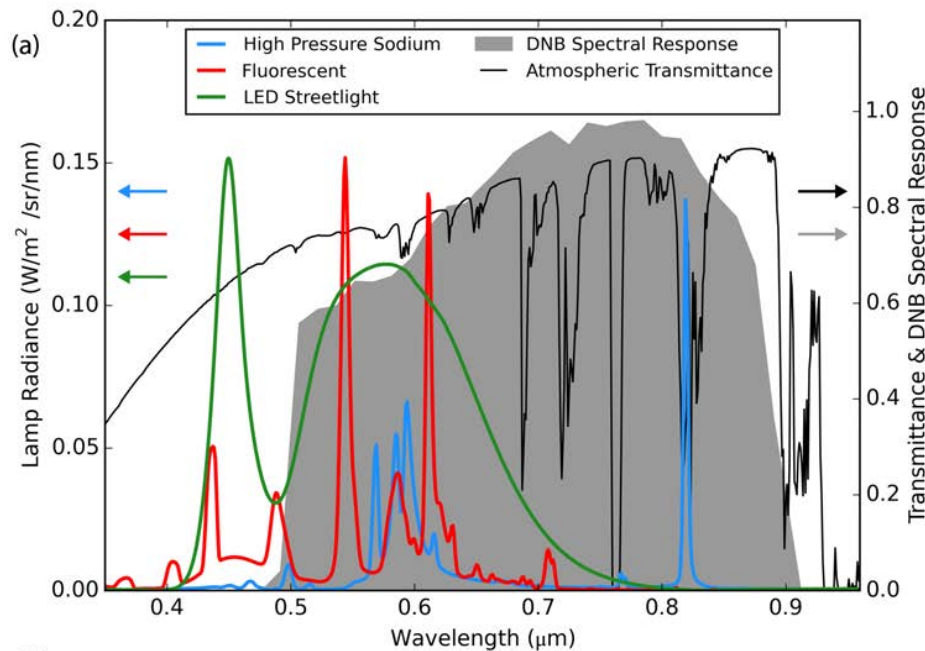


Miller et al., 2013



Wang et al., 2014

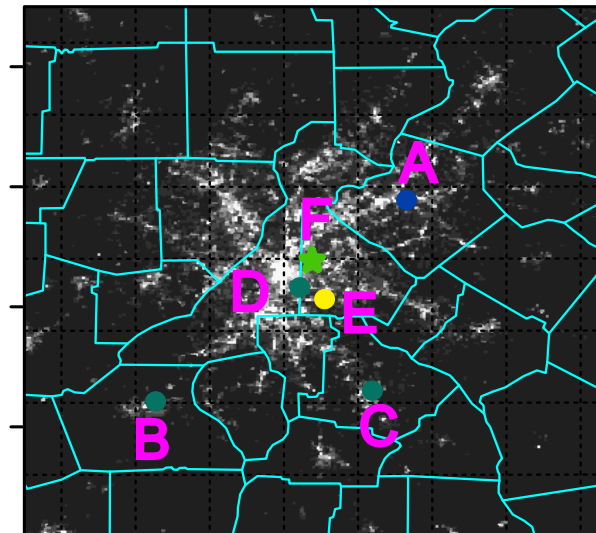
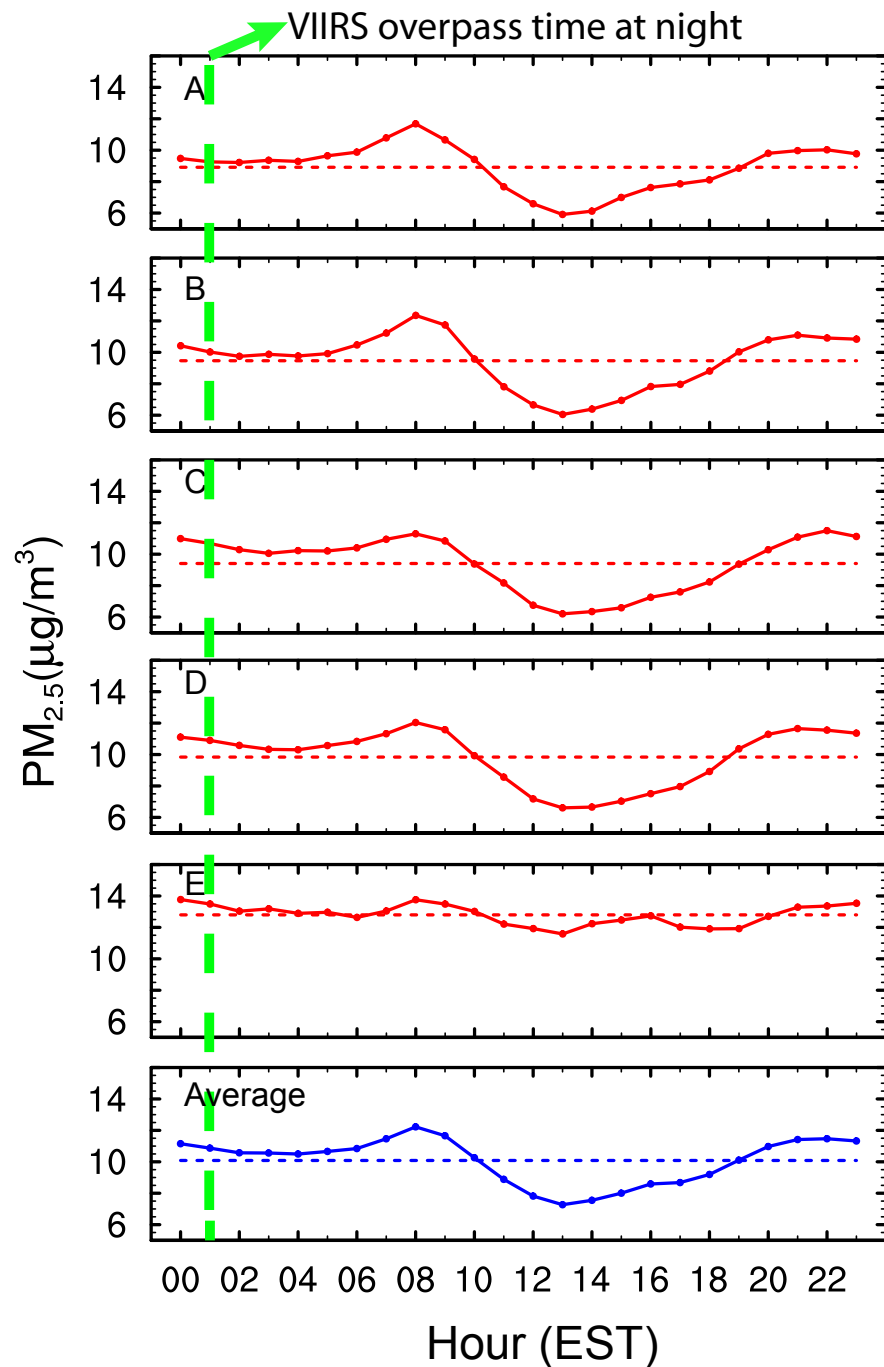
DNB is most sensitive to change of AOD but, water vapor effect is also not negligible



The database of spectral intensity emitted from HPS, fluorescent, and LED bulbs are from Elvidge et al., (2010).

In the U.S., high- pressure sodium lamps (HPS) are the most common type of light source used for outdoor applications (Rea et al., 2009)

**PM_{2.5} at VIIRS night
overpass time is more
representative daily-mean
PM_{2.5} than at noon time
(VIIRS daytime overpass)**



Regression analysis

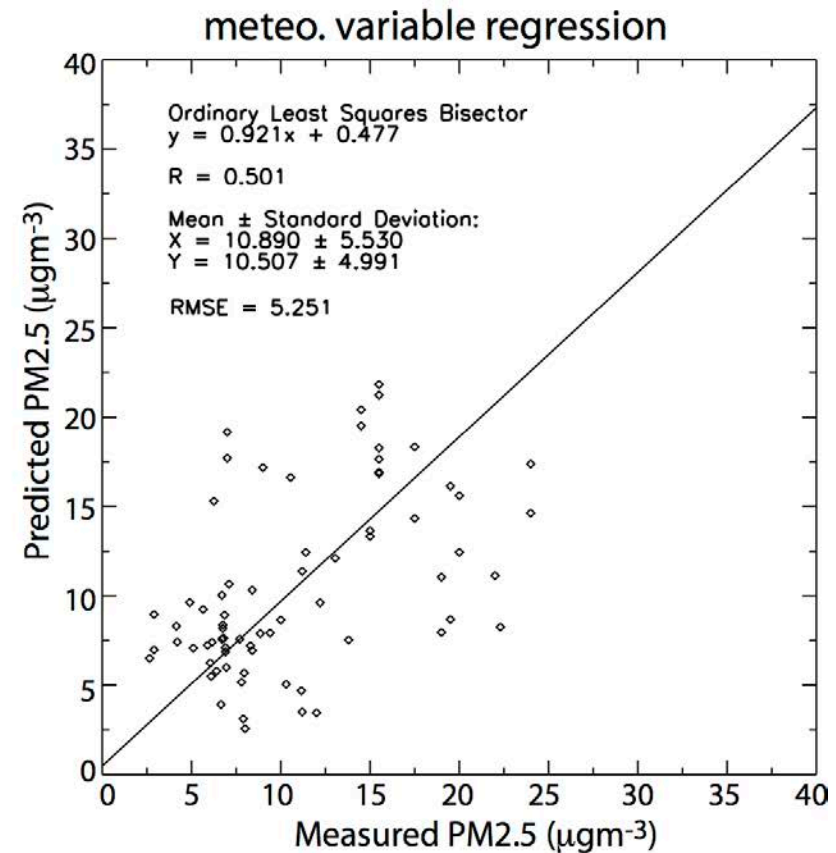
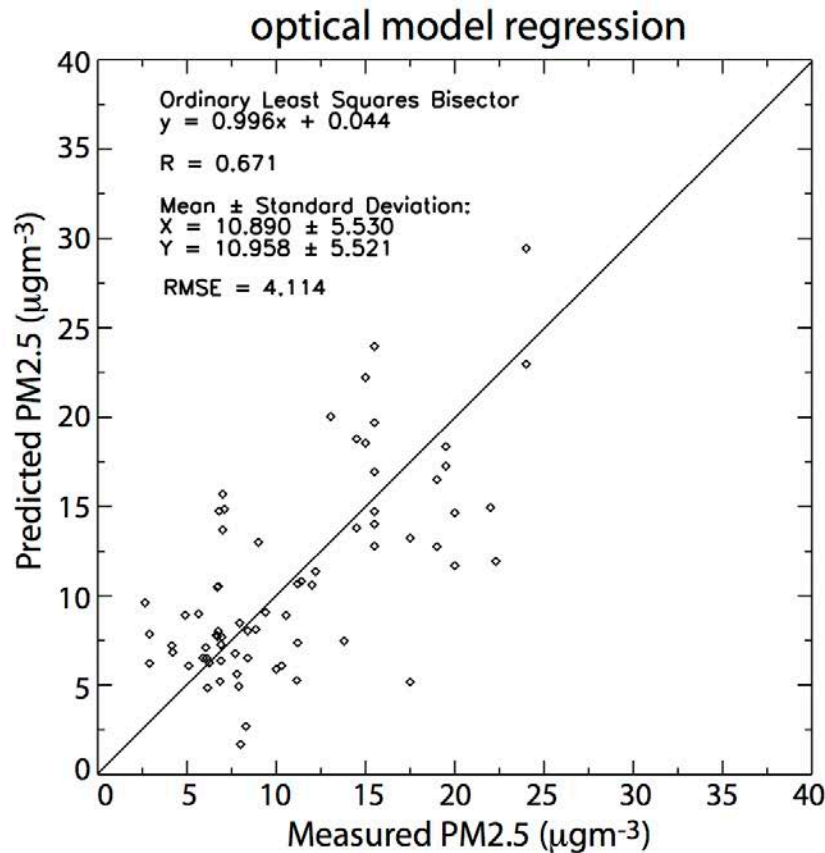
Among surface wind speed, surface pressure, and columnar water vapor amount) that are routinely measured at the surface, the DNB light intensity is the only variable that shows either the largest or second largest correlation with surface PM_{2.5}.

Table 2. Correlation coefficients (R) between PM_{2.5}×f(rh)/μ and different variables at 6 ground sites (A-F as described in Table 1 and marked in Figure 2)¹.

	A	B	C	D	E	F(CTR)
Variables\R						
lnI	-0.78	-0.56	<i>-0.53</i>	<i>-0.39</i>	-0.71	-0.73
ΔPs	0.05	0.21	0.08	0.14	0.10	0.10
W	0.49	0.38	0.85	0.17	0.00	0.10
U×f(rh)/μ	-0.21	-0.08	-0.21	-0.30	<i>-0.60</i>	<i>-0.66</i>
V×f(rh)/μ	<i>0.59</i>	<i>0.49</i>	0.48	0.53	0.54	0.52

¹At each site, the largest value is in bold and second largest value is in the italic bold

Leave-one-out cross validation of regression model



$$f(rh)/\mu = a_0 - a_1 \ln(I) - a_2 r \times W - a_3 p \times P_s$$

$$PM_{2.5} = f(W, P_s, U, V)$$

VIIRS-based optical model gives better estimate of surface $PM_{2.5}$ than meteorology-based regression.

Improving Nocturnal Fire Detection With the VIIRS Day–Night Band

Thomas N. Polivka, Jun Wang, Luke T. Ellison, Edward J. Hyer, and Charles M. Ichoku

Abstract—Building on existing techniques for remotely sensing fires via satellite, this paper takes advantage of the day–night band (DNB) aboard the Visible Infrared Imaging Radiometer Suite (VIIRS) to develop the Firelight Detection Algorithm (FILDA), which characterizes fire pixels based on both visible-light and infrared (IR) signatures at night. By adjusting fire pixel selection criteria to include visible-light signatures, FILDA allows for significantly improved detection of pixels with smaller and/or cooler subpixel hotspots than the operational Interface Data Processing System (IDPS) algorithm. VIIRS scenes with near-coincident Advanced Spaceborne Thermal Emission and Reflection (ASTER) overpasses are examined after applying the operational VIIRS fire product algorithm and including a modified “candidate fire pixel selection” approach from FILDA that lowers the $4\text{-}\mu\text{m}$ brightness temperature (BT) threshold but includes a minimum DNB radiance. FILDA is shown to be effective in detecting gas flares and characterizing fire lines during large forest fires (such as the Rim Fire in California and High Park fire in Colorado). Compared with the operational VIIRS fire algorithm for the study period, FILDA shows a large increase (up to 90%) in the number of detected fire pixels that can be verified with the finer resolution ASTER data (90 m). Part (30%) of this increase is likely due to a combined use of DNB and lower $4\text{-}\mu\text{m}$ BT thresholds for fire detection in FILDA. Although further studies are needed, quantitative use of the DNB to improve fire detection could lead to reduced response times to wildfires and better estimate of fire characteristics (smoldering and flaming) at night.

that, despite improving warning systems [1], [2], have exacted 38 greater costs in recent years [3], [4]. In addition, they im- 39 pact global atmospheric chemistry by releasing potent trace 40 gases such as carbon monoxide, carbon dioxide, methane, and 41 ethene [5], as well as aerosols and black carbon [6]. These by- 42 products of combustion are capable of traveling great distances 43 and impacting health and meteorological processes in remote 44 locations [7], [8], and in addition to creating local pollution 45 hazards, these can affect Earth’s climate [9]. Fire-spawned 46 smoke aerosols have complex interactions with the atmosphere 47 by causing a reduction in surface illumination [10]–[12] and 48 simultaneously warming the atmosphere, thereby decreasing 49 vertical temperature gradients and increasing atmospheric sta- 50 bility [13] due to their relatively low single-scattering albedo 51 [14]. As a consequence of wildfire lethality and potential 52 for property damage, earlier detection of wildfires via remote 53 sensing is paramount to proper allocation of fire manage- 54 ment resources [15], [16]. Effective response to all of these 55 phenomena requires accurately detecting and characterizing 56 fires as well as accurately quantifying emissions from biomass 57 burning. 58

The launch of the Suomi National Polar-orbiting Partnership 59 (SNPP) satellite on 28 October 2011 has opened up unprece- 60

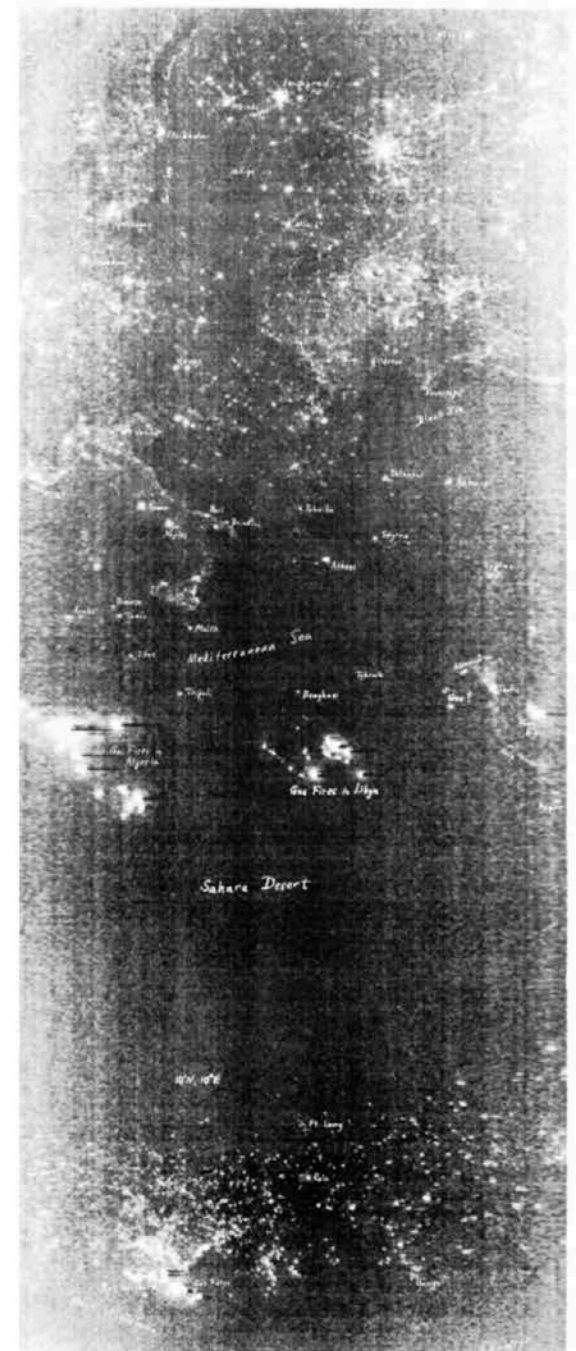
First fire detection from space was from visible light at night...

Burning Waste Gas in Oil Fields

I WAS recently amazed by some night-time spacecraft photographs, exemplified by Fig. 1, that present graphic evidence of waste and pollution. These were obtained by the United States Air Force DAPP system which has sensors in the visible 0.4 to 1.1 μm band and an infrared imaging system in the 8 to

- T. A. Croft, *Nature*, 1973.

Such agricultural “Fires, invisible by day, are seen ranging all around ... at night (when) we were literally surrounded by them; some smouldering, ... others fitfully bursting forth, whilst others again stalked along with a steadily increasing and enlarging flame...” Hooker (in 1846), cited by Croft, 1973.



Nighttime Images of the Earth from Space

An unusual aspect of the earth is revealed in pictures recorded at midnight by U.S. Air Force weather satellites. The brightest lights on the dark side of the planet are giant waste-gas flares

by Thomas A. Croft

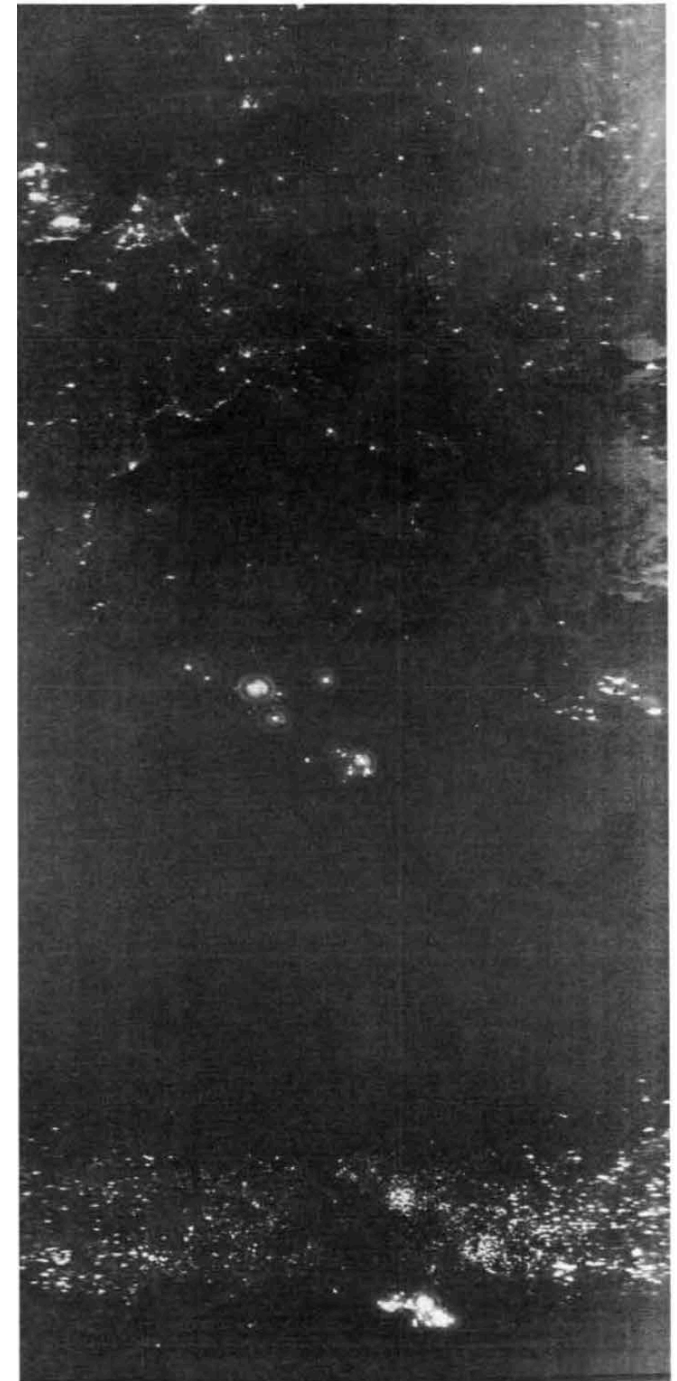
THREE MAJOR LIGHT SOURCES associated with human activities are visible in this nighttime satellite image ...

the upper third of this picture are the **city lights** of Europe.

The larger isolated lights near the middle and bottom arise from **gas flares** at oil fields in Al-geria, Libya and Nigeria.

The uniform band of smaller lights scattered across Africa south of the Sahara appears to originate with **agricultural and pastoral fires**.

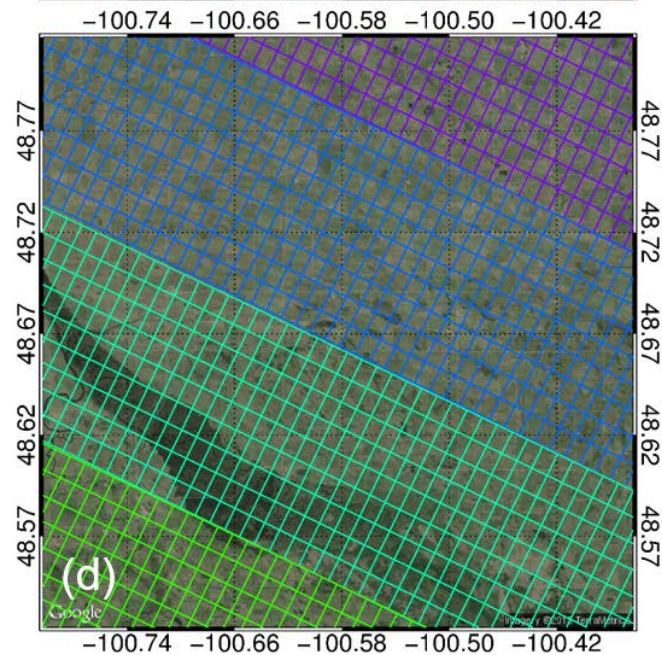
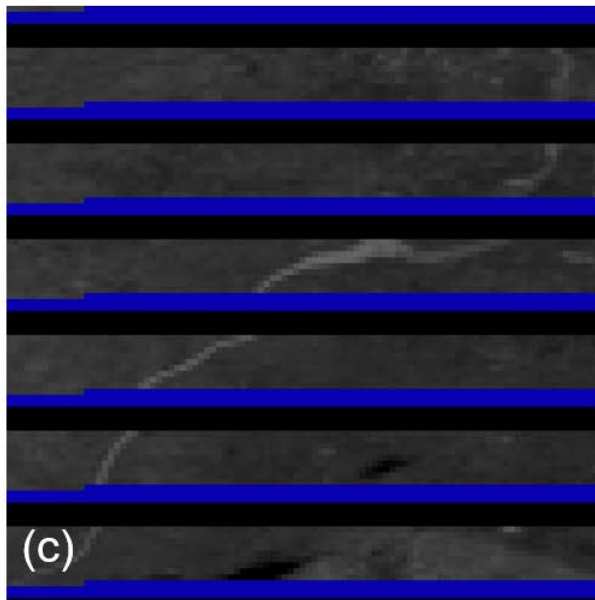
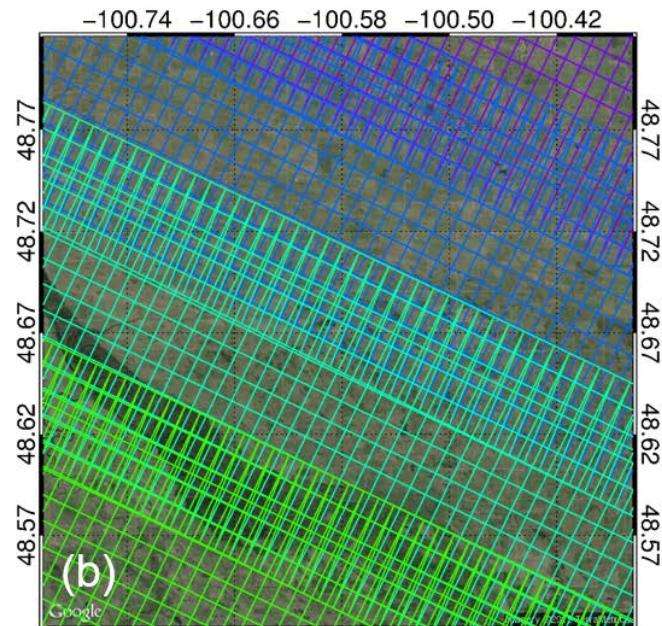
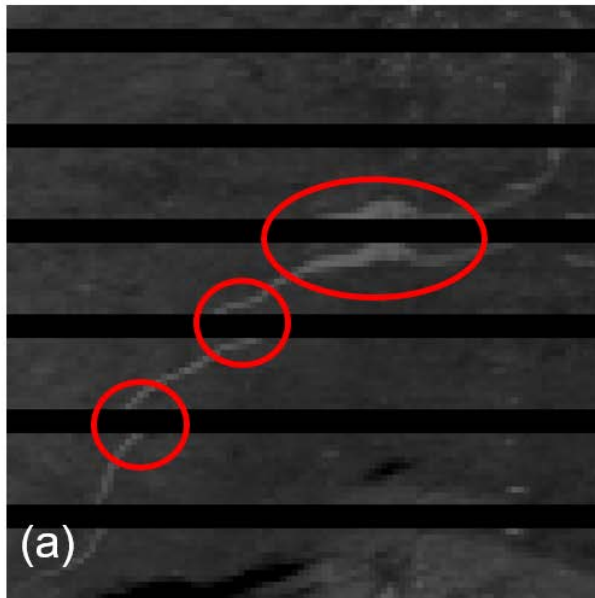
Scientific America, 1978.



Recent work of using shortwave IR (1.6 μm) for night fire detection

- C. D. Elvidge, M. Zhizhin, F.-C. Hsu, and K. E. Baugh, “VIIRS nightfire: 1333 Satellite pyrometry at night,” *Remote Sens.*, vol. 5, no. 9, pp. 4423–4449, 1334 Sep. 2013, doi: 10.3390/rs5094423.
- W. Schroeder, P. Oliva, L. Giglio, and I. A. Csiszar, “The new VIIRS 375 m active fire detection data product: Algorithm description and initial assessment,” *Remote Sens. Environ.*, vol. 143, pp. 85–96, Mar. 5, 2014, doi: 10.1016/j.rse.2013.12.008.

Correction of pixel overlaps

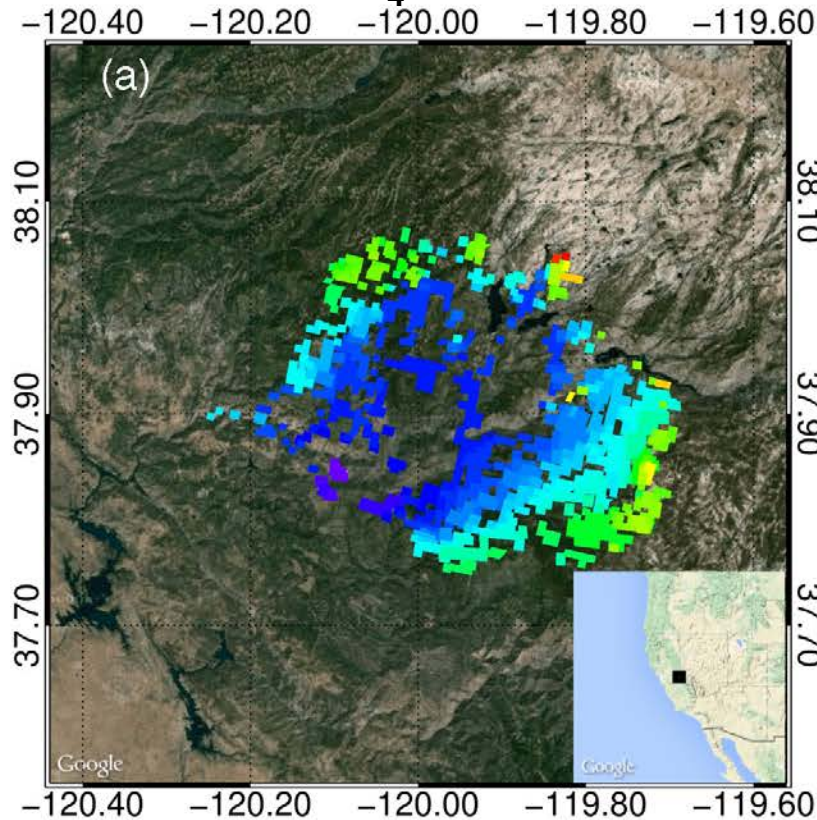


M13

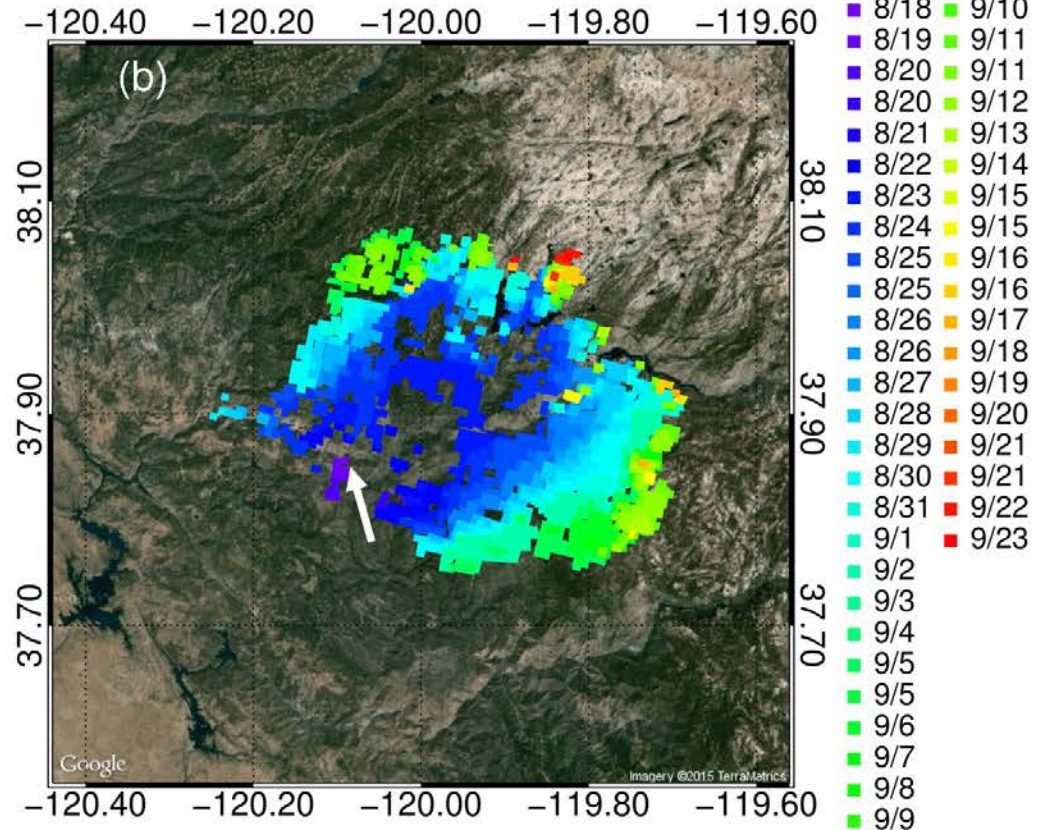
Firelight Detection Algorithm (FILDA)

Combined use of Vis + NIR + IR to detect fires

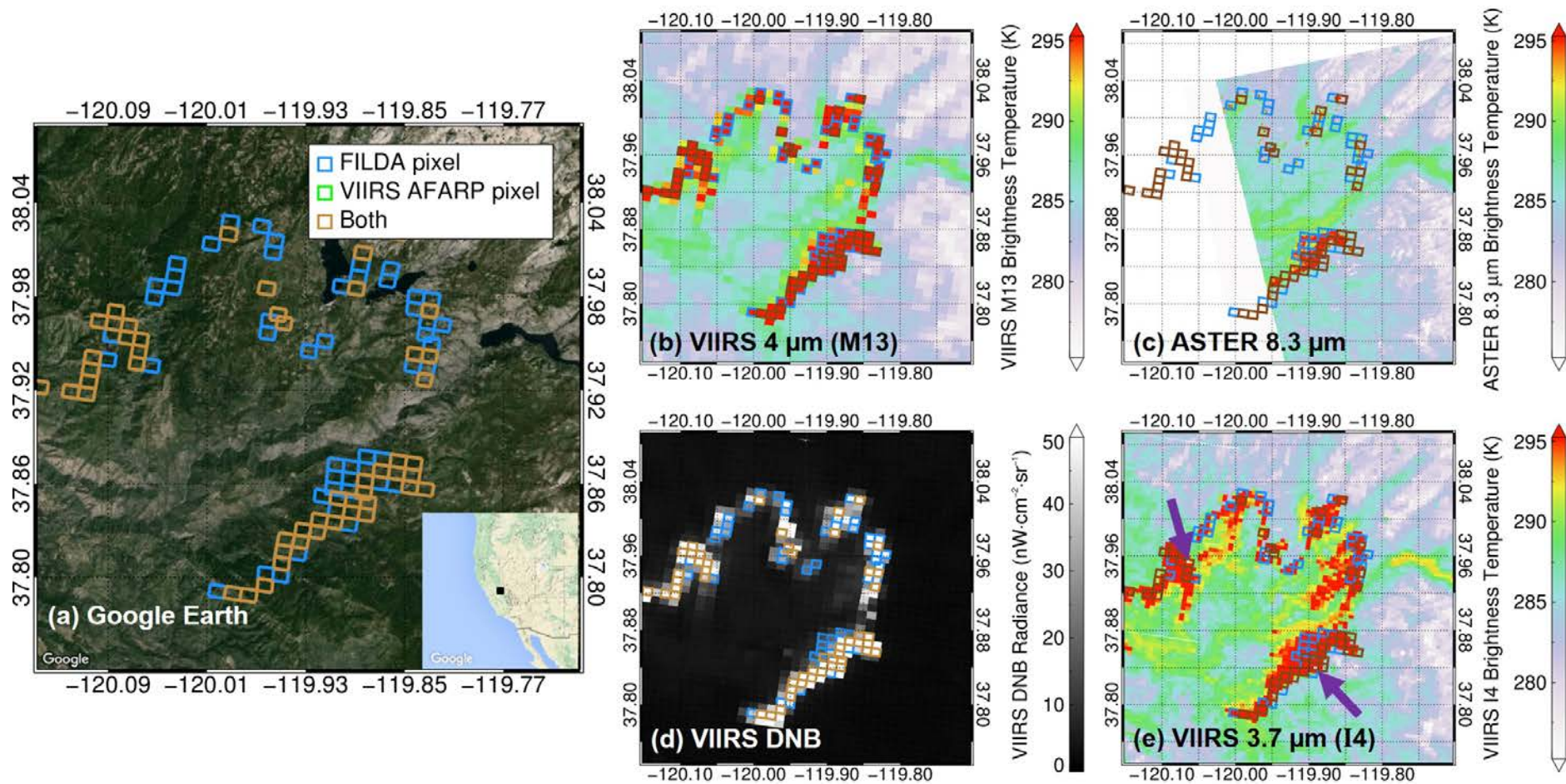
IDPS AFARP
Active Fire Application Related Product
 $BT_4 > 320$ K



FILDA
Dynamic threshold of BT_4
& DNB threshold

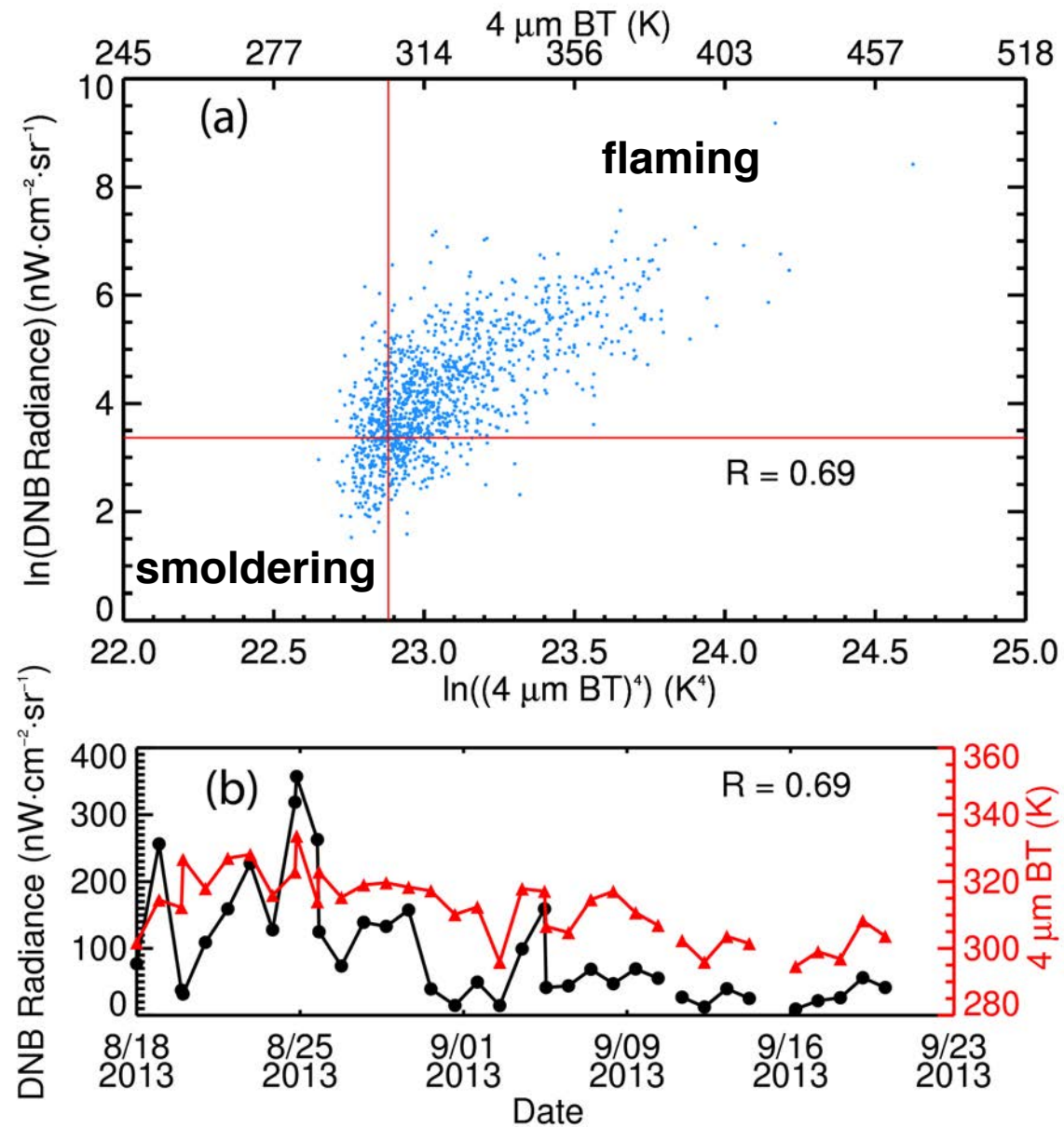


Evaluation with ASTER

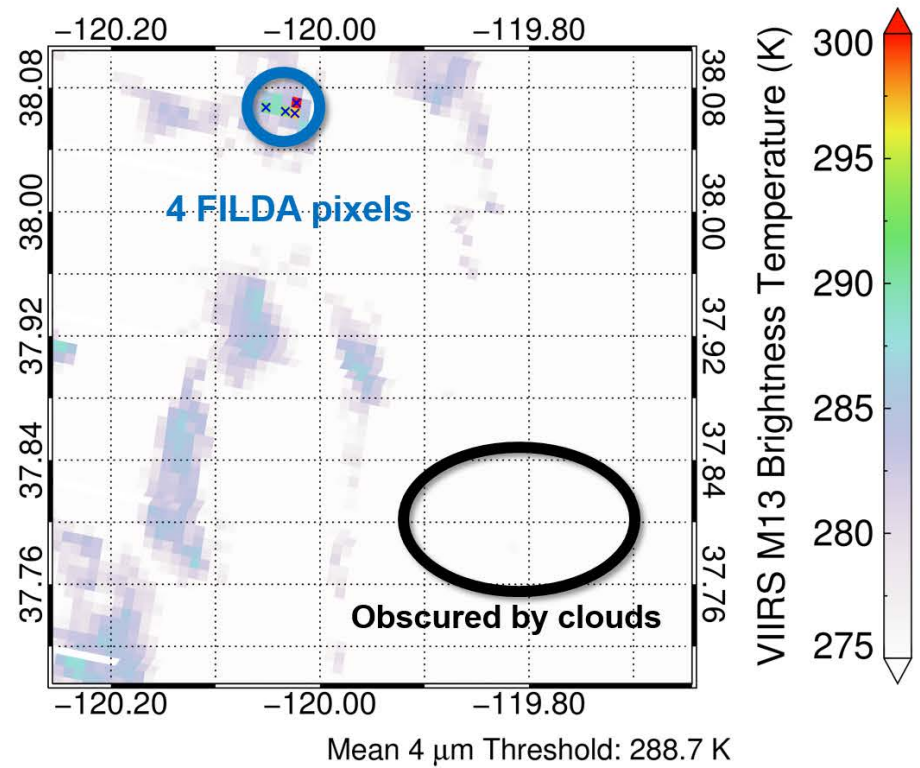
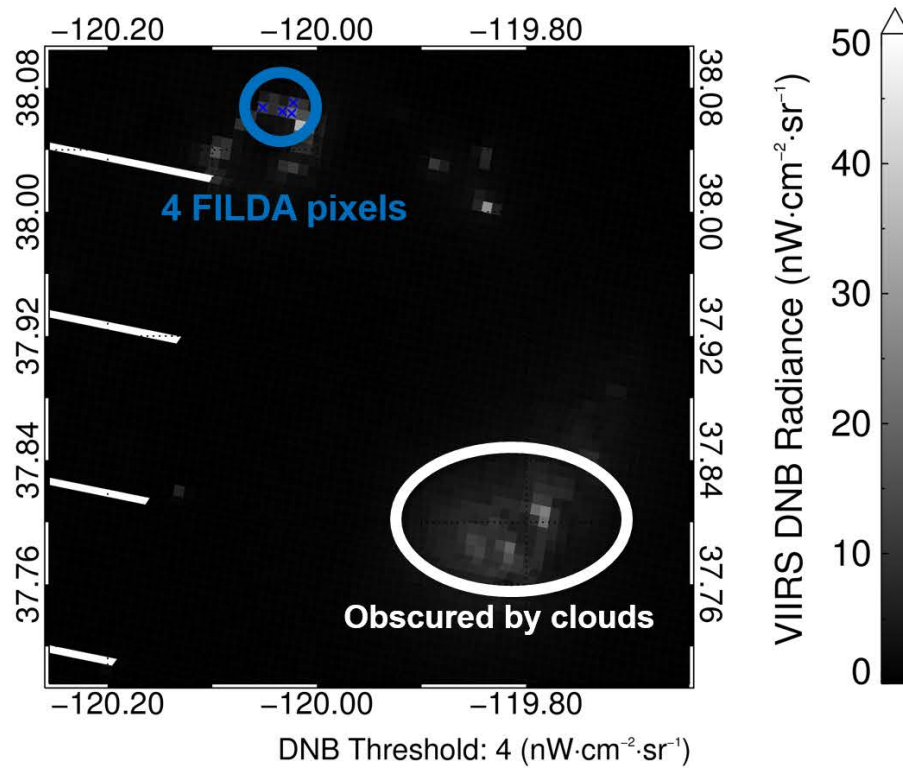


Multiband/sensor view of the Rim Fire taken at 2:29 AM PDT, 24 August 2013

Potential characterization of smoldering vs. flaming



Potential to detect fires through clouds



Thank you !

Acknowledgement:

NASA Suomi-NPP program & NASA Applied Science program.

Polivka, T., J. Wang, L. Ellison, E. Hyer, and C. Ichoku, Improving Nocturnal Fire Detection with the VIIRS Day-Night Band, *IEEE Transactions on Geoscience & Remote Sensing*, in press, 2016.

Wang, J., Clint Aegerter, Xiaoguang Xu, and J. J. Szykman, Potential application of VIIRS Day/Night Band for monitoring nighttime surface PM_{2.5} air quality from space, *Atmospheric Environment*, 124, 55–63, doi:10.1016/j.atmosenv.2015.11.013, 2016.

Polivka, T., E. Hyer, J. Wang, and D. Peterson, First global analysis of saturation artifacts in the VIIRS infrared channels and the effects of sample aggregation, *IEEE Geoscience and Remote Sensing Letters*, 1262-1266, 2015, (journal cover article).



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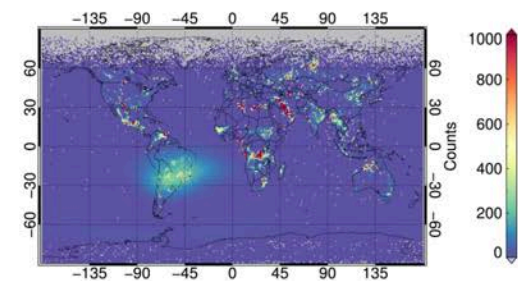
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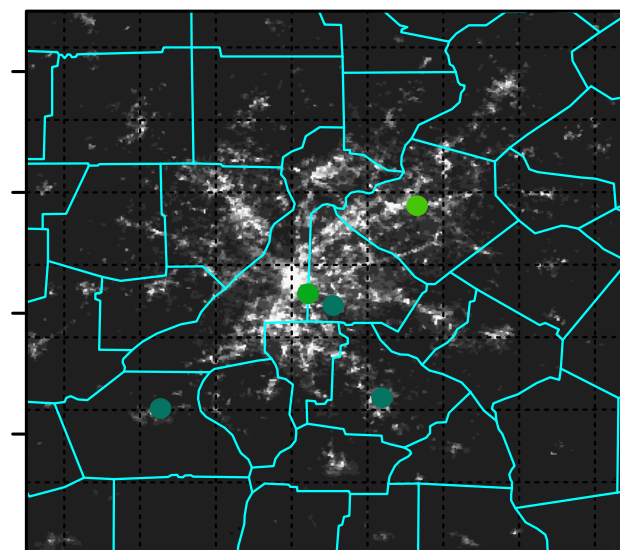
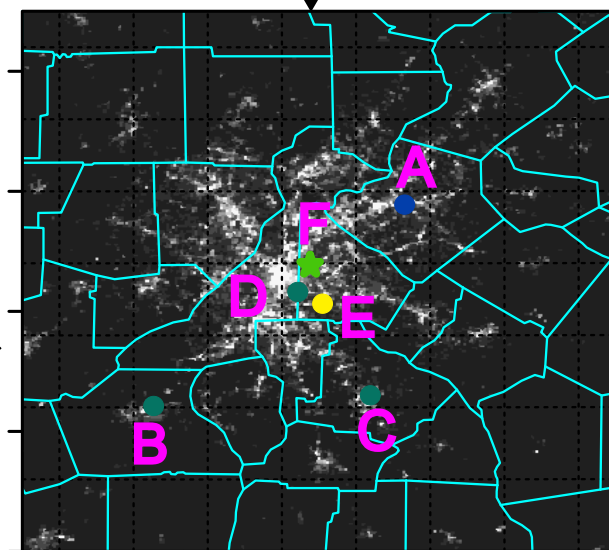
Regions of frequent biomass burning and gas flaring highlighted by NOAA's Nightfire product during 18 March–14 July 2013. Interference from the South Atlantic Anomaly (SAA) is visible over a large portion of South America.

Case Analysis

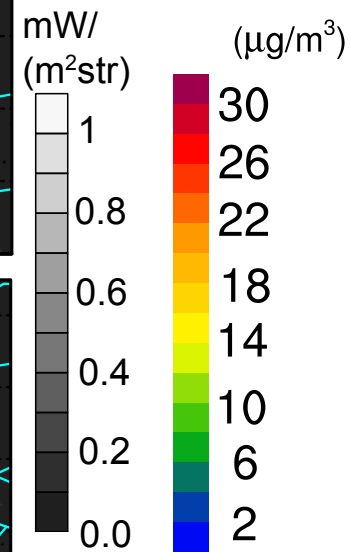
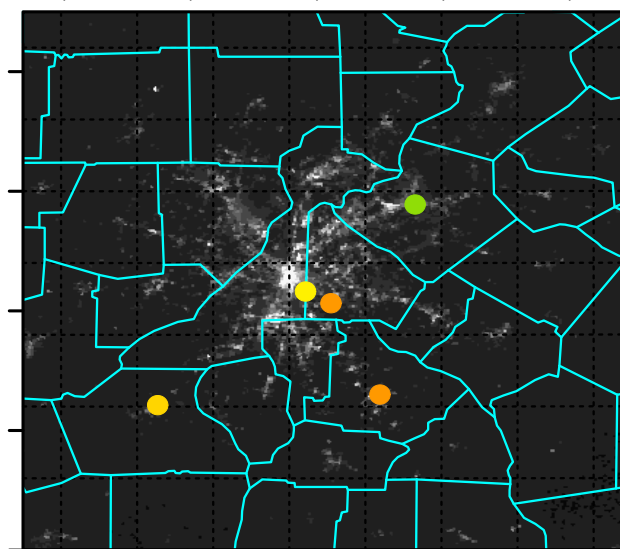
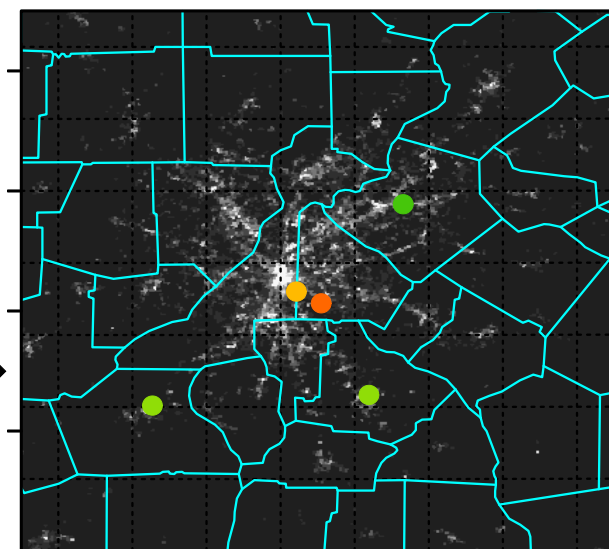
Moon Nights

Moonless Nights

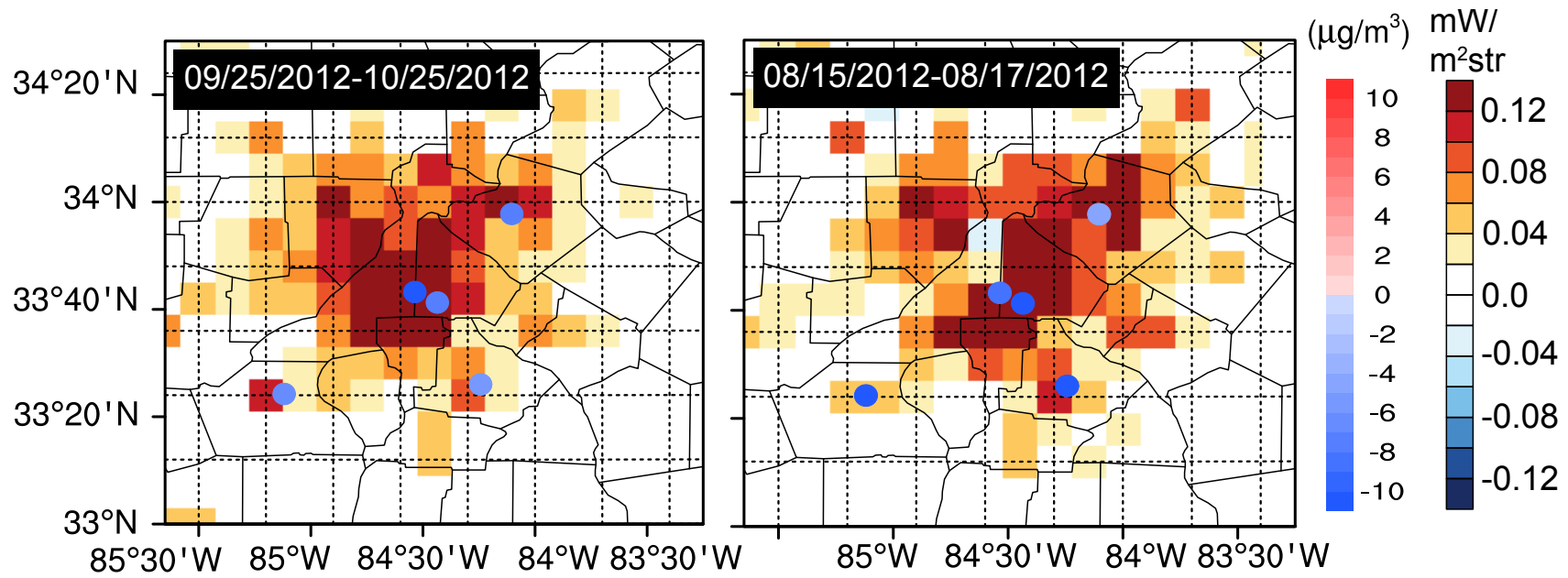
Low $\text{PM}_{2.5}$
Large VZA →



High $\text{PM}_{2.5}$
Nadir view →



City lights enhance the signal for detecting PM_{2.5} changes.



Using moonlight alone to detect PM_{2.5} appears very challenging.